

# K-Nearest Neighbors

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Nipun Batra

IIT Gandhinagar

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**High values of K** will result in smoother decision boundaries

$\Rightarrow$  lower variance but also higher bias

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What is the time complexity of finding  $k$  nearest neighbors naively?



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- **Scalability:** Approximate methods needed for large datasets